APPLICATION
FOR
UNITED STATES LETTERS PATENT

TO WHOM IT MAY CONCERN:

Be it know that we, Shai Dekel and Nitzan Goldberg have invented new and useful improvements in a

SYSTEM AND METHOD FOR THE LOSSLESS PROGRESSIVE STREAMING OF IMAGES OVER A COMMUNICATION NETWORK

for which the following is a specification.

5 **SYSTEM AND METHOD FOR THE LOSSLESS PROGRESSIVE STREAMING OF IMAGES OVER A COMMUNICATION NETWORK**

This application claims priority to U.S. Utility Patent Application No.
10 09/386,264 filed August 31, 1999, entitled "SYSTEM AND METHOD FOR TRANSMITTING A DIGITAL IMAGE OVER A COMMUNICATION NETWORK", and U.S. Provisional Patent Application No. 60/198,017, filed April 18, 2000, entitled "LOSSLESS PROGRESSIVE STREAMING OF IMAGES OVER THE INTERNET", the entirety of which are both incorporated herein by reference.

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BACKGROUND OF THE INVENTION

FIELD OF THE INVENTION
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This invention relates to systems and methods for transmission of still images over relatively low-speed communication channels. More specifically the invention relates to progressive image streaming over low speed communication lines, and may be applied to a variety of fields and disciplines, including commercial printing and
25 medical imaging, among others.

BRIEF DESCRIPTION OF THE PRIOR ART

30 In a narrow bandwidth environment, a simple transfer to the client computer of any original image stored in the server's storage is obviously time consuming. In many cases the user only wishes to view a low resolution of the image and perhaps a few more high-resolution details, in these instances it would be inefficient to transfer the full image. This problem can be overcome by storing images in some compressed
35 formats. Examples for such formats are standards such as Progressive JPEG (W. Pennebaker and J. Mitchel, "JPEG, still image data compression standard", VNR, 1993) or the upcoming JPEG2000 (D. Taubman, "High performance scalable image compression with EBCOT", preprint, 1999). These formats allow progressive transmission of an image such that the quality of the image displayed at the client
40 computer improves during the transmission.

5   In some application such as medical imaging, it is also necessary that whenever the user at the client computer is viewing a portion of the highest resolution of the image, the progressive streaming will terminate at **lossless** quality. This means that at the end of progressive transmission the pixels rendered on the screen are exactly the pixels of the original image. The current known "state-of-the-art" wavelet

10   algorithms for progressive lossless streaming all have a major drawback: their rate-distortion behavior is very inferior to the "lossy" algorithms. The implications are serious:

1. Whenever the user is viewing any low resolution of the image (at low
15   resolutions the term "lossless" is not well defined) more data needs to be sent for the same visual quality.

2. During the progressive transmission of the highest resolution, before lossless quality is achieved, more data needs to be sent for the same visual quality.

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Researchers working in this field are troubled by these phenomena. As F. Sheng, A. Bilgin, J. Sementilli and M. W. Marcellin say in [SBSM]: "…Improved lossy performance when using integer transforms is a pursuit of our on-going work." Here is an example:

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Wavelet	Rate (bit per pixel)				
	0.1	0.2	0.5	0.7	1.0
Floating Point 7x9	24.18	26.65	31.64	34.17	36.90
Reversible (4,4)	23.89	26.41	31.14	33.35	35.65

Table 1 – Comparison of the lossy compression performances (implemented by the $(7,9)$ Wavelet) to a lossless compression (implemented by a reversible $(4,4)$ Wavelet) of "Barabara" image
30   **(PSNR (dB)) ([SBSM]).**

5   As one can see from Table 1, state of the art progressive lossless coding is inferior to lossy coding by more than 1 dB at the high bit-rate.

Indeed, intuitively, the requirement for lossless progressive image transmission should not effect the rendering of lower resolutions or the progressive "lossy" rendering of the highest resolution before lossless quality is obtained. The final lossless quality

10   should be a layer that in some sense is added to a lossy algorithm with minor (if any) effect on its performance.

The main problem with known lossless wavelet algorithms, such as SPIHT [SP1] and CREW [ZASB], is that they use special "Integer To Integer" transforms (see "Wavelet transforms that map integers to integers", A. Calderbank, I. Daubechies, W.

15   Sweldens, B. L. Yeo, J. Fourier Anal. Appl., 1998). These transforms mimic "mathematically proven" transforms that work well in lossy compression using floating-point arithmetic implementations. But because they are constraint to be lossless, they do not approximate their floating-point ancestors sufficiently well. Although in all previous work there have been attempts to correct this approximation in

20   the progressive coding stage of the algorithm, the bad starting point, an inefficient transform, prevented previous authors to obtain decent rate-distortion behavior.

Our algorithm solves the rate-distortion behavior problem. Using the fact that images are two-dimensional signals, we introduce new 2D lossless Wavelet transforms that approximate much better their lossy counterparts. As an immediate consequence

25   our lossless progressive coding algorithm has the same rate-distortion of a lossy algorithm during the lossy part of the progressive transmission.

SUMMARY OF THE INVENTION

The imaging system that is described below is directed to a lossless image

30   streaming system that is different from traditional compression systems and overcomes the above problems. By utilizing a lossless means of progressive transmission means the pixels rendered on the screen at the end of transmission are exactly the pixels of the original image that was transmitted. The imaging system disclosed herein eliminates the necessity to store a compressed version of the original image, by streaming ROI

35   data using the original stored image. The imaging system of the present invention also avoids the computationally intensive task of compression of the full image. Instead,

5   once a user wishes to interact with a remote image, the imaging server performs a fast

preprocessing step in near real time after which it can respond to any ROI requests also

in near real time. When a ROI request arrives at the server, a sophisticated progressive

image-encoding algorithm is performed, but not for the full image. Instead, the

encoding algorithm is performed only for the ROI. Since the size of the ROI is bounded

10   by the size and resolution of the viewing device at the client and not by the size of the

image, only a small portion of the full progressive coding computation is performed for

a local area of the original image. This local property is also true for the client. The

client computer performs decoding and rendering only for ROI and not the full image.

This real time streaming or Pixels-On-Demand™ architecture requires different

15   approaches even to old ideas. For example, similarly to some prior art, the present

imaging system is based on wavelets. But while in other systems wavelet bases are

selected according to their coding abilities, the choice of wavelet bases in the present

imaging system depends more on their ability to perform well in the real time

framework. The system of the present invention supports several modes of progressive

20   transmission: by resolution, by accuracy and by spatial order.

BRIEF DESCRIPTION OF THE DRAWINGS

Figure 1 is a system architecture block diagram.

25   Figure 2 is an imaging system workflow diagram.

Figure 3 is a flow diagram representing a "lossless progressive by accuracy"

request list for a ROI.

Figure 4 is a diagram depicting the client "progressive by accuracy" workflow.

Figure 5 is a diagram depicting the server workflow.

30   Figure 6 is a diagram describing the server-preprocessing step.

Figure 7 is a diagram describing the low resolution encoding process.

Figure 8 is a diagram describing the ROI high resolution processing.

Figure 9 is a diagram depicting the local forward lossless wavelet transform.

Figure 10 is a diagram depicting the local inverse lossless wavelet transform.

35   Figure 11 is a diagram depicting the progressive rendering steps.

5   Figure 12 is a diagram depicting a lossless subband tile wherein the spatial grouping of subband coefficients are at a given resolution and the halfbit matrix is associated with the hh_j subband.

Figure 13 a diagram depicting the RGB <-> YUV reversible conversion.

Figure 14 a diagram depicting a lossless last bit plane data block in which only

10   hl and hh subband coefficients are scanned.

Figure 15 is a sample pseudo-code of the encoding algorithm represented by: (a) Least significant bit plane scan pseudo-code (b) Half bit plane scan pseudo-code.

Figure 16 is a sample pseudo-code of the decoding algorithm represented by: (a) Least significant bit plane scan pseudo-code (b) Half bit plane scan pseudo-code.

15   Figure 17 is a diagram depicting the curve defining the mapping from Original_Image_Depth-bit image to Screen_Depth-bit.

Figure 18 is a diagram depicting the decomposition of one-dimensional signal x to the Low-subband s and the High-subband d and the separable decomposition of two-dimensional signal X into 4 matrices (subbands): LL, HL, LH and HH.

20   Figure 19 is a diagram depicting the first stage of the 2D separable transform step in the X-direction.

Figure 20 is a diagram depicting the second stage of the 2D separable transform step in the Y-direction.

Figure 21 is a diagram depicting the application of the full 2D Wavelet

25   transform.

Figure 22 is a flow diagram representing the least significant bit plane scan of the encoding algorithm.

Figure 23 is a flow diagram representing the least significant bit plane scan of the decoding algorithm.

30   Figure 24 is a flow diagram describing a bit plane significance scan of the server-encoding algorithm.

Figure 25 is a flow diagram describing the subband tile extraction of the client progressive rendering process.

Figure 26 is a diagram depicting the preprocessing multi-resolution structure.

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DETAILED DESCRIPTION OF THE INVENTION

1.   NOTATION AND TERMINOLOGY

The following notation is used throughout this document.

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Term	Definition
4D	Four dimensional
FIR	Finite Impulse Response
FWT	Forward Wavelet Transform
GUI	Graphical User Interface
ID	Identification tag
IWT	Inverse Wavelet Transform
ROI	Region Of Interest
URL	Uniform Resource Locator
LSB	Least Significant Bit
RMS	Root Square Error
FP	Floating Point
PDF	Probability Distribution Function

The following terminology and definitions apply throughout this document.

Term	Definition
Rendering	Procedure to output/display a ROI of an image into a device such as a monitor, printer, etc.
Multiresolution	A pyramidal structure representing an image at dyadic resolutions, beginning with the original image as the highest resolution.
Subband Transform / subband coefficients	A method of decomposing an image (time domain) to frequency components (frequency domain). A representation of an image as a sum of differences between the dyadic resolutions of the image's multiresolution.
Wavelet Transform / Wavelet coefficients	A special case of Subband Transform.
Progressive Transmission	Transmitting a given image in successive steps, where each step adds more detail to the rendering of the image
Progressive Rendering	A sequence of rendering operations, each adding more detail.
Progressive by accuracy	Transmit/render strong features first (sharp edges), less significant features (texture) last
Progressive by resolution	Transmit/render low resolution first, high resolution last
Progressive by spatial order	Transmit/render top of image first, bottom of image last
Distributed database	A database architecture that can be used in a network environment

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Subband/Wavelet tile	A group of subband/wavelet coefficients corresponding to a time-frequency localization at some given spatial location and resolution/frequency
Subband/Wavelet data block	An encoded portion of a subband/wavelet tile that corresponds to some accuracy layer

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2. OVERVIEW OF THE INVENTION

Referring to Figure 1, a block diagram is provided depicting the various components of the imaging system in one embodiment. A client computer 110 is coupled to a server computer 120 through a communication network 130.

10   In one embodiment, the client computer 110 and server computer 120 may comprise a PC-type computer operating with a Pentium-class microprocessor, or equivalent. Each of the computers 110 and 120 may include a cache 111, 121 respectively as part of its memory. The server may include a suitable storage device 122, such as a high-capacity disk, CD-ROM, DVD, or the like.

15   The client computer 110 and server computer 120 may be connected to each other, and to other computers, through a communication network 130, which may be the Internet, an Intranet (e.g., a local area network), a wide-area network, or the like. Those having ordinary skill in the art will recognize that any of a variety of communication networks may be used to implement the present invention.

20   With reference to Figure 2, the system workflow is described. Using any browser type application, the user of the client computer 110 connects to the Web Server 140 or directly to the Imaging server 120 as described in Figure 1. He/she then selects, using common browser tools an image residing on the Image file storage 122. The corresponding URL request is received and processed by the Imaging Server 120.

25   In case results of previous computations previous computations on the image are not present in the Imaging Cache 121, the server performs a fast preprocessing algorithm (see §2.1) in a lossless mode. The result of this computation is inserted into the cache 121. Unlike other more "traditional" applications or methods which perform full progressive encoding of the image in an "offline" type method, the goal of the

30   preprocessing step is to allow the server, after a relatively fast computational step, to serve any ROI specified by the user of the client computer. For example, for a 15M grayscale medical image, using the described (software) server, installed on computer

5   with a Pentium processor, fast disk, running Windows NT, the preprocessing step 501

will typically take 3 seconds. This is by order of magnitude faster than a "full"

compression algorithm such as [S], [SP1], [T]. Serving ROI is sometimes called

"pixels-on-demand" which means a progressive transmission of any ROI of the image

in "real-time", where the quality of the view improves with the transfer rate until

10   lossless view is received in the client side. Once the preprocessing stage is done, the

server sends to the client a notification that the "image is ready to be served". The

server also transmits the basic parameters associated with the image such as

dimensions, color space, etc. Upon receiving this notification, the client can select any

ROI of the image using standard GUI. The ROI is formulated in step 203 into a request

15   list that is sent to the server. Each such request corresponds to a data block (§4). The

order of requests in the list corresponds to some progressive mode selected in the

context of the application such as "progressive by accuracy" rendering of the ROI.

Upon receiving the ROI request list, the server processes the requests according to their

order. For each such request the server checks if the corresponding data block exists in

20   the Cache 121. If not, the server then computes the data block, stores it in the cache and

immediately sends it to the client. Once a data block that was requested arrives at the

client, it is inserted into the cache 111. At various points in time during the transfer

process, a decision rule invokes a rendering of the ROI. Obviously, if some of the data

blocks required for a high quality rendering of the ROI, were requested from the server,

25   but have not arrived yet, the rendering of the ROI will be of lower quality. But, due to

the progressive request order, the rendering quality will improve with each received

data block in an "optimal" way. In case the user changes the ROI, the rendering task at

the client is canceled and a new request list corresponding to a new ROI, sent from the

client to the server, will notify the server to terminate the previous computation and

30   transmission task and begin the new one.

3.   NEW REVERSIBLE WAVELET TRANSFORM

We will now explain in detail why the rate-distortion behavior of our

35   progressive lossless algorithm is better than other known algorithms. Lossless wavelet

transform, must be integer-to-integer transform, such that round-off errors are avoided.

5   In order to demonstrate the difference between lossy and lossless transforms, let us look

at the simplest wavelet, the Haar wavelet. Let $x(k)$ be the k-th component of the one-

dimensional discrete signal x. The first forward Haar transform step, in its accurate

"mathematical" form, is defined by:

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$$\begin{cases} s(n) = \dfrac{1}{\sqrt{2}}\left(x(2n+1)+x(2n)\right), \\ d(n) = \dfrac{1}{\sqrt{2}}\left(x(2n+1)-x(2n)\right), \end{cases} \tag{3.1}$$

where s is a low-resolution version of x, and d is the "difference" between s and x.

In the case of lossless transform, applying the above transform results in round-off

error. One possibility is to apply the transform step suggested by [CDSI]:

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$$\begin{cases} s(n) = \left\lfloor \dfrac{x(2n+1)+x(2n)}{2} \right\rfloor, \\ d(n) = x(2n+1)-x(2n). \end{cases} \tag{3.2}$$

The symbol $\lfloor \circ \rfloor$ is the floor function meaning "greatest integer less than or equal to $\circ$",

20   e.g.

$$\lfloor 0.5 \rfloor = 0, \lfloor -1.5 \rfloor = -1, \lfloor 2 \rfloor = 2, \lfloor -1 \rfloor = -1.$$

The one-dimensional transform step is generalized to 2D separable transform

step, by applying the 1D transform step twice, first in the X-direction and than (on the

25   first stage output) in the Y-direction as described in Figure 18, 19 and 20 (see also [M]

7.7). The full 2D Wavelet transform is applied by using the 2D Wavelet transform step

iteratively in the classic Mallat decomposition of the image (Figure 21) ([M] 7.7).

In (3.2) two properties are kept:

30   1. **Reversibility**, i.e., one can restore $x(2n)$ and $x(2n+1)$, by knowing $s(n)$ and

$d(n)$, as follows:

$$\begin{cases} x(2n) = s(n) - \left\lfloor \dfrac{d(n)}{2} \right\rfloor, \\ x(2n+1) = d(n) + x(2n). \end{cases} \tag{3.3}$$

2. **De-correlation**, i.e., $s(n)$ and $d(n)$ contains the minimal number of bits required in order to follow property 1. For example, if the transform would have been defined by:

$$\begin{cases} s(n) = x(2n+1) + x(2n), \\ d(n) = x(2n+1) - x(2n), \end{cases} \tag{3.4}$$

then the least significant bit of $s(n)$ and $d(n)$ would have been the same and saved twice. In other words, there is a correlation between $s(n)$ and $d(n)$ in (3.4). From the view point of coding this should be avoided since there is a redundancy of transmitting this bit.

On the other hand, the **scaling** property, that is a very important one, is not kept in (3.2). Observe that the value of $s(n)$ computed by (3.2), is smaller than its "real mathematical value" as computed in (3.1), by factor of $\sqrt{2}$. Since $s(n)$ should be rounded to an integer number, the fact that $s(n)$ is smaller than what it should be, increases the round-off error. In low resolutions, the error is accumulated through the wavelet steps.

If we take the error as a model of "white noise" added to the $i-th$ resolution in a multi-resolution representation of the image, i.e. X_i in Figure 21, it can be proved that the variance of this noise exponentially increases as a function of i. This "contamination" to the multi-resolution image damages the coding efficiency at low bit-rates. Let us describe this in detail for the case of the Haar wavelet. We have two assumptions in our analysis:

1. The parity (least significant bit) of an arbitrary coefficient c, in any of the wavelet steps is a uniformly distributed random variable, i.e.

$$\Pr(c \equiv 0 \mod 2) = \Pr(c \equiv 1 \mod 2) = \frac{1}{2}.$$

5   2. This parity is independent of other coefficient's parity (Identically independent distributed).

Our referenced computation, i.e. the accurate computation, is the Haar transform step defined in (3.1). We concentrate on the LL-subband coefficients, because the low-

10   resolution subbands are computed from them. LL-subband coefficients are the result of a 2D-transform step (Figure 18)

$$ll^{(\text{accurate})}(m,n) = \frac{x(2m+1,2n+1)+x(2m,2n+1)+x(2m+1,2n)+x(2m,2n)}{2},\cdot$$

where m and n are the indices of the row and column of the coefficient respectively.

15   As described in Figure 18, 19 and 20, according to the step defined in (3.2), we first apply the X-direction step

$$s(k,n) = \left\lfloor \frac{x(k,2n+1)+x(k,2n)}{2} \right\rfloor,$$

for each input row $x(k,\cdot)$.

20   Under assumption 1 mentioned above, we can write $s(k,n)$ as

$$s(k,n) = \frac{x(k,2n+1)+x(k,2n)}{2} + e, \tag{3.5}$$

25   where e is a random variable with a probability distribution function (P.D.F.) $p(\cdot)$ defined by

$$\begin{cases} p(-0.5) = \Pr(e=-0.5) = \dfrac{1}{2}, \\ p(0) = \Pr(e=0) = \dfrac{1}{2}. \end{cases} \tag{3.6}$$

Therefore,

30   $$\mathrm{E}(e) = -\frac{1}{4}, \mathrm{Var}(e) = \frac{1}{16}. \tag{3.7}$$

5   We then apply the Y-direction transform by

$$ll^{(\text{CDSI})}(m,n) = \left\lfloor \frac{s(2m+1,n)+s(2m,n)}{2} \right\rfloor = \frac{s(2m+1,n)+s(2m,n)}{2}+e'. \qquad (3.8)$$

As in (3.5) we can represent $s(2m+1,n)$ and $s(2m,n)$ by:

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$$s(2m+1,n) = \frac{x(2m+1,2n+1)+x(2m+1,2n)}{2}+e_1, \qquad (3.9)$$

$$s(2m,n) = \frac{x(2m,2n+1)+x(2m,2n)}{2}+e_2. \qquad (3.10)$$

Now we can write:

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$$ll^{(\text{CDSI})}(m,n)$$

$$= \frac{\dfrac{x(2m+1,2n+1)+x(2m+1,2n)}{2}+e_1+\dfrac{x(2m,2n+1)+x(2m,2n)}{2}+e_2}{2}+e'$$

$$(3.11)$$

$$= \frac{x(2m+1,2n+1)+x(2m+1,2n)+x(2m,2n+1)+x(2m,2n)}{4}+\frac{e_1}{2}+\frac{e_2}{2}+e'$$

$$= \frac{x(2m+1,2n+1)+x(2m+1,2n)+x(2m,2n+1)+x(2m,2n)}{4}+e$$

where e_1, e_2, e' are independent (assumption 2 above) random variables with

20   expectation $\dfrac{1}{4}$, Variance $\dfrac{1}{16}$, and $e = \dfrac{e_1}{2}+\dfrac{e_2}{2}+e'$.

Therefore,

$$E(e) = \frac{1}{2}E(e_1) + \frac{1}{2}E(e_2) + E(e') = -\frac{1}{2},$$

$$\text{Var}(e) = \text{Var}\left(\frac{e_1}{2} + \frac{e_2}{2} + e'\right) = \frac{1}{4}\text{Var}(e_1) + \frac{1}{4}\text{Var}(e_2) + \text{Var}(e') = \quad (3.12)$$

$$= \frac{1}{4} \cdot \frac{1}{16} + \frac{1}{4} \cdot \frac{1}{16} + \frac{1}{16} = \frac{3}{32}.$$

Thus,

$$ll^{(\text{CDSI})}(m,n) = \frac{ll^{(\text{accurate})}(m,n)}{2} + e. \quad (3.13)$$

e represents the approximation error of the LL-subband coefficients, results from one 2D transform step. The error relates to the accurate floating-point computation.

This was a description of a single 2D-transform step assuming that the input coefficients are without any error. Now we wish to evaluate the error accumulated after several steps.

At an arbitrary step $i \geq 0$, we can assume that an input coefficient can be written as:

$$x_i(k,l) = x_i^{(\text{accurate})}(k,l) + e_i,$$

where $x_i^{(\text{accurate})}(k,l)$ is the accurate value achieved by floating-point computation for all the previous steps, i.e., a step defined by

$$\begin{cases} s(n) = \dfrac{x(2n+1) + x(2n)}{2}, \\ d(n) = x(2n+1) - x(2n), \end{cases} \quad (3.14)$$

instead of the integer-to-integer computation in (3.2). Observe that if $x_i^{(\text{accurate})}(k,l)$ is the i-th resolution image coefficient, using (3.14) as the 1D Wavelet step, then

$$x_i^{(\text{accurate})}(k,l) = \frac{ll_{i-1}^{(\text{accurate})}(k,l)}{2^i} \quad , i \geq 1, \quad (3.15)$$

where $ll_{i-1}^{(\text{accurate})}(k,l)$ is the normalized ($L_2 - norm$) LL-subband coefficient resulting from the i-th 2D transform step using (3.1) as the 1D Wavelet step (see Figure). e_i is

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5 the difference between $x_i(k,l)$ and $x_i^{(\text{accurate})}(k,l)$ (I.e., the approximation error of the

integer computation made until now). E.g. $e_0 = 0$ ($x_0(k,l)$ is an original image pixel),

while e_1 is a random number with expectation $-\dfrac{1}{2}$ and variance $\dfrac{3}{32}$ (see (3.12)).

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Using (3.11), we get:

$$ll_i^{(\text{CDSI})}(m,n)$$

$$= \frac{x_i^{(\text{acc})}(2m+1,2n+1)+e_i^1+x_i^{(\text{acc})}(2m+1,2n)+e_i^2+x_i^{(\text{acc})}(2m,2n+1)+e_i^3+x_i^{(\text{acc})}(2m,2n)+e_i^4}{4}+e$$

$$= \frac{x_i^{(\text{acc})}(2m+1,2n+1)+x_i^{(\text{acc})}(2m+1,2n)+x_i^{(\text{acc})}(2m,2n+1)+x_i^{(\text{acc})}(2m,2n)}{4}+\frac{e_i^1+e_i^2+e_i^3+e_i^4}{4}+e$$

$$= \frac{x_i^{(\text{acc})}(2m+1,2n+1)+x_i^{(\text{acc})}(2m+1,2n)+x_i^{(\text{acc})}(2m,2n+1)+x_i^{(\text{acc})}(2m,2n)}{4}+e_{i+1}$$

$$= x_{i+1}^{(\text{accurate})}+e_{i+1},$$

where e_{i+1} is defined by

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$$e_{i+1} = \frac{e_i^1+e_i^2+e_i^3+e_i^4}{4}+e,$$

and corresponds to LL_i subband.

Consequently

$$\mathrm{E}(e_{i+1}) = \mathrm{E}(e_i)+\mathrm{E}(e),$$
$$\mathrm{Var}(e_{i+1}) = \frac{\mathrm{Var}(e_i)}{4}+\mathrm{Var}(e).$$

Observe that

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$$\mathrm{E}(e)=-\frac{1}{2},\mathrm{Var}(e)=\frac{3}{32}.$$

As a result, we can write recursive formulas for the error expectation and variance after
i steps.

$$\begin{cases} \mathrm{E}\left(e_0\right) = 0, \\ \mathrm{E}\left(e_{i+1}\right) = \mathrm{E}\left(e_i\right) - \dfrac{1}{2}, \end{cases}$$

$$\begin{cases} \mathrm{Var}\left(e_0\right) = 0, \\ \mathrm{Var}\left(e_{i+1}\right) = \dfrac{\mathrm{Var}\left(e_i\right)}{4} + \dfrac{3}{32}, \end{cases} \tag{3.16}$$

The explicit solutions to these formulas are

$$\mathrm{E}\left(e_i\right) = -\frac{i}{2}, \mathrm{Var}\left(e_i\right) = \frac{1}{8} - \frac{1}{2\cdot 4^{i+1}}. \tag{3.17}$$

By replacing $x_i^{(\mathrm{accurate})}\left(m,n\right)$ with $\dfrac{ll_{i-1}^{(\mathrm{accurate})}\left(m,n\right)}{2^i}$ we get

$$ll_i^{(\mathrm{CDSI})}\left(m,n\right) = \frac{ll_i^{(\mathrm{accurate})}\left(m,n\right)}{2^{i+1}} + e_{i+1}. \tag{3.18}$$

Thus, the approximation to $ll_i^{(\mathrm{accurate})}\left(m,n\right)$ is

$$2^{i+1} ll_i^{(\mathrm{CDSI})}\left(m,n\right) = ll_i^{(\mathrm{accurate})}\left(m,n\right) + 2^{i+1} e_{i+1}.$$

The approximation error expectation is

$$\mathrm{E}\left(2^i e_i\right) = 2^i \mathrm{E}\left(e_i\right) = 2^i\left(-\frac{i}{2}\right) = -i\, 2^{i-1}.$$

The approximation error variance and standard deviation are

$$\mathrm{Var}\left(2^i e_i\right) = 4^i \mathrm{Var}\left(e_i\right) = 4^i\left(\frac{1}{8} - \frac{1}{2\cdot 4^{i+1}}\right) = \frac{4^i - 1}{8} \approx \frac{4^i}{8} = 2^{2i-3}.$$

Hence

$$\mathrm{Std}\left(2^i e_i\right) = \sqrt{\mathrm{Var}\left(2^i e_i\right)} \approx \frac{2^{i-1}}{\sqrt{2}}.$$

5   Let us now evaluate the approximation error of the 3 other subbands:

$$lh_i^{(CDSI)}(m,n) = \left\lfloor \frac{ll_{i-1}^{(CDSI)}(2m+1,2n+1) + ll_{i-1}^{(CDSI)}(2m+1,2n)}{2} \right\rfloor - \left\lfloor \frac{ll_{i-1}^{(CDSI)}(2m,2n+1) + ll_{i-1}^{(CDSI)}(2m,2n)}{2} \right\rfloor$$

$$= \frac{lh_i^{(accurate)}(m,n)}{2^i} + \frac{e_i^1 + e_i^2}{2} + e' - \left(\frac{e_i^3 + e_i^4}{2} + e'' \right) = \frac{lh_i^{(accurate)}(m,n)}{2^i} + e_i^{LH}$$

where

10   • e_i^k $1 \leq k \leq 4$ are identical to the random variable whose expectation and variance are given in (3.17).

   • $e_i^{LH} = \frac{e_i^1 + e_i^2}{2} + e' - \left(\frac{e_i^3 + e_i^4}{2} + e'' \right)$

   • e' and e'' are identical to the random variable whose expectation and variance are given in (3.7).

15   Thus,

$$\mathrm{E}\left(e_i^{LH}\right) = 0,$$

$$\mathrm{Var}\left(e_i^{LH}\right) \approx \frac{1}{4}\left(\frac{1}{8} - \frac{1}{2 \cdot 4^{i+1}} + \frac{1}{8} - \frac{1}{2 \cdot 4^{i+1}} \right) + \frac{1}{16} \qquad (3.19)$$

$$+ \frac{1}{4}\left(\frac{1}{8} - \frac{1}{2 \cdot 4^{i+1}} + \frac{1}{8} - \frac{1}{2 \cdot 4^{i+1}} \right) + \frac{1}{16}$$

$$= \frac{1}{4} - \frac{1}{2 \cdot 4^{i+1}}.$$

The approximation to $lh_i^{(accurate)}(m,n)$ is

20          $$2^i lh_i^{(CDSI)}(m,n) = lh_i^{(accurate)}(m,n) + 2^i e_i^{LH}.$$

The approximation error variance and standard deviation are:

$$\mathrm{Var}\left(2^i e_i^{LH}\right) = 4^i \mathrm{Var}\left(e_i^{LH}\right) = 4^i \cdot \left(\frac{1}{4} - \frac{1}{2 \cdot 4^{i+1}} \right) = 4^{i-1} - \frac{1}{8} \approx 4^{i-1}.$$

Therefore

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5

$$\text{Std}\left(2^i e_i^{LH}\right) = \sqrt{\text{Var}\left(2^i e_i^{LH}\right)} \approx \sqrt{4^{i-1}} = 2^{i-1}.$$

A similar approximation error estimation, it can be done with the HL-subband and the HH-subband.

10 The approximation error evaluation results are summarized in the following table where the error is the difference between the normalized (in L_2-norm) coefficients according to [CDSI] reversible transform and the "mathematical" transform (defined in (3.1)).

	Expectation	Variance	Std
LL_i - error	$-(i+1)2^i$	$\dfrac{4^{i+1}-1}{8} \approx 2^{2i-1}$	$0.707 \cdot 2^i$
LH_i - error	0	$\dfrac{2 \cdot 4^i - 1}{8} \approx 4^{i-1}$	$0.5 \cdot 2^i$
HL_i - error	$-\dfrac{1}{4} \cdot 2^i$	$\dfrac{3 \cdot 4^i - 2}{16} \approx 3 \cdot 4^{i-2}$	$0.433 \cdot 2^i$
HH_i - error	0	$\dfrac{4^i-1}{8} \approx 2^{2i-3}$	$0.354 \cdot 2^i$

15 **Table 2 - Normalized (in L_2-norm) approximation errors of the Wavelet coefficients at resolution $i \geq 0$ ($i = 0$ is the highest resolution) using the (CDSI) reversible Haar transform.**

Assuming a low-bit rate transmission where only the coefficients whose absolute
20 value belongs to the range $\left[2^b, 2^{b+1}\right)$ are encoded, for every resolution i, where i is greater than b (less or more). It must be noted that the large error implies a significant loss of coding efficiency.

Instead, we propose a new family of reversible transforms. The proposed family of integer wavelet transforms has all three properties:

25 1. **Reversibility**

2. **De-correlation**

3. **Scaling – i.e. improved approximation of the "mathematical" transform.**

Our 2D transform step is separable also, but the one-dimensional transform step, which the 2D transform is based on, is different for the X-direction (step 1901), the Y-direction step applied on the low output of the X-direction step (step 2001) and the Y-direction step applied on the high output of the X-direction step (step 2002) as described in Figure 18, 19 and 20.

The full 2D Wavelet transform is applied by using the 2D Wavelet transform step iteratively in the classic Mallat decomposition of the image (Figure 21) ([M] 7.7). As mentioned before, the Wavelet coefficients in our proposed transform are all scaled, i.e. normalized in $L_2-\text{norm}$ as the Wavelet coefficients computed in the accurate "mathematical" transform.

In order to achieve the third property (improved approximation of the "mathematical" transform), we define an extra matrix we call the "Half bit-matrix" which enables the reversibility of the High Y-transform step (step 2002). The elements that belong to this matrix are bits, such that each bit corresponds to an HH-subband coefficient in the following interpretation. Let us describe this by the following example.

Supposing

$$s(n) = 7, d^{(1)}(n) = 9$$

are a coefficient pair results from a reversible de-correlated 1D-wavelet step

$$\begin{cases} d^{(1)}(n) = x(2n+1) - x(2n), \\ s(n) = \left\lfloor \dfrac{x(2n) + x(2n+1)}{2} \right\rfloor. \end{cases}$$

Now, $d^{(1)}(n)$ has to be multiplied by $\dfrac{1}{2}$, in order to be scaled.

The binary form of $d^{(1)}(n) = 9$ is

$$d^{(1)}(n) = 1001_2.$$

If we now divide $d^{(1)}(n)$ by 2 in a floating-point computation we get

$$d^{FP}(n) = \frac{1}{2}d^{(1)}(n) = 100.1_2.$$

Let us call the bit, which located on the write-side of the floating point the "Half Bit". Observe that the Half Bit of $d^{FP}(n)$ is the LSB of $d^{(1)}(n)$. Therefore, an equivalent way to do this in an integer computation without loosing the Half-Bit is to calculate first the LSB of $d^{(1)}(n)$ by

$$HalfBit(n) = d^{(1)}(n) \bmod 2 = 9 \bmod 2 = 1,$$

then to shift-write $d^{(1)}(n)$ by

$$d(n) = d^{(1)}(n) \gg 1 = 1001 \gg 1 = 100.$$

By saving $d(n)$ and $HalfBit(n)$ we can restore back $d^{(1)}(n)$.

In the proposed transform, this Half-bit is needed in the HH-subband coefficient computation. Therefore in our wavelet decomposition for every HH-subband coefficient (in all scales) there is a corresponding bit, which is the coefficient's Half-bit.

The Half bit matrix is hidden in the HH-subband in the description of Figure 18, 19 and 20. It is described explicitly in the specification of the transform as much in the coding algorithm.

We now present our integer-to-integer versions of the Haar transform and the CDF (1,3) transform for the 2-dimensional case.

3.1 Reversible Haar and (CDF) (1,3) Transforms

3.1.1 Haar Transform

With respect to Figure 19:

3.1.1.1 Step 1901: X-direction

Forward Step

$$\begin{cases} s(n) = \left\lfloor \dfrac{x(2n) + x(2n+1)}{2} \right\rfloor, \\ d(n) = x(2n+1) - x(2n). \end{cases} \tag{3.20}$$

Inverse Step

$$\begin{cases} x(2n) = s(n) - \left\lfloor \dfrac{d(n)}{2} \right\rfloor, \\ x(2n+1) = d(n) + x(2n). \end{cases} \qquad (3.21)$$

With respect to Figure 20:

3.1.1.2 Step 2001: Y-direction – Low Forward Step

$$\begin{cases} s(n) = x(2n) + x(2n+1), \\ d^{(1)}(n) = \left\lfloor \dfrac{x(2n+1) - x(2n)}{2} \right\rfloor, \\ d(n) = 2d^{(1)}(n). \end{cases} \qquad (3.22)$$

Remarks:

1. $s(n)$ is a scaled LL-subband coefficient.

2. $s(n)$ and $d^{(1)}(n)$ are de-correlated and a reversible couple (can be transformed back to $x(2n)$ and $x(2n+1)$), but $d^{(1)}(n)$ is not scaled (it is half its "real value"). Thus, $d^{(1)}(n)$ is multiplied by 2. Nevertheless, the LSB of the LH-subband coefficient $d(n)$ is known to be 0 and not encoded.

Inverse Step

$$\begin{cases} x(2n+1) = \dfrac{1}{2}\left(s(n) + d(n) + (s(n)\bmod 2)\right), \\ x(2n) = s(n) - x(2n+1). \end{cases} \qquad (3.23)$$

With respect to Figure 20:

5   **3.1.1.3 Step 2002: Y-direction – High Forward Step**

$$\begin{cases} d^{(1)}(n) = x(2n+1) - x(2n), \\ HalfBit(n) = \left(d^{(1)}(n)\right)\bmod 2, \\ d(n) = \left\lfloor \dfrac{d^{(1)}(n)}{2} \right\rfloor, \\ s(n) = x(2n) + d(n). \end{cases} \qquad (3.24)$$

Remark: $d^{(1)}(n)$ and $s(n)$ are de-correlated and reversible couples, but $d^{(1)}(n)$ is not

scaled (It is twice its "real value"). Therefore, $d^{(1)}(n)$ is divided by 2. By doing that,

we lose its least significant bit, which cannot be restored. To solve this problem, as

10   explained before, we save this bit as the "**Half-Bit**". Giving this name to that

coefficient means that its weight is $\dfrac{1}{2}$ in the "real mathematical scale", and it is the

least significant (from the approximation point of view).

      Inverse Step

$$\begin{cases} x(2n) = s(n) - d(n), \\ d^{(1)}(n) = 2d(n) + HalfBit(n), \\ x(2n+1) = d^{(1)}(n) + x(2n). \end{cases} \qquad (3.25)$$

15

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5   **3.1.2 CDF (1,3) Transform**

3.1.2.1 Step 1901: X-direction

With respect to Figure 19:

Forward Step

10
$$\begin{cases} s(n) = \left\lfloor \dfrac{x(2n)+x(2n+1)}{2} \right\rfloor, \\[2ex] d^{(1)}(n) = x(2n+1) - x(2n), \\[2ex] d(n) = d^{(1)}(n) + \left\lfloor \dfrac{s(n-1)-s(n+1)}{4} \right\rfloor. \end{cases} \tag{3.26}$$

Inverse Step

$$\begin{cases} d^{(1)}(n) = d(n) - \left\lfloor \dfrac{s(n-1)-s(n+1)}{4} \right\rfloor, \\[2ex] x(2n) = s(n) - \left\lfloor \dfrac{d^{(1)}(n)}{2} \right\rfloor, \\[2ex] x(2n+1) = x(2n) + d^{(1)}(n). \end{cases} \tag{3.27}$$

With respect to Figure 20:

3.1.2.2 Step 2001: Y-direction – Low Forward Step

15
$$\begin{cases} s(n) = x(2n) + x(2n+1), \\[2ex] d^{(1)}(n) = \left\lfloor \dfrac{x(2n+1)-x(2n) + \left\lfloor \dfrac{s(n-1)-s(n+1)}{8} \right\rfloor}{2} \right\rfloor, \\[2ex] d(n) = 2d^{(1)}(n). \end{cases} \tag{3.28}$$

Remark: See remarks for (3.22).

22

5

Inverse Step

$$\begin{cases} s^{(1)}(n) = s(n) - \left\lfloor \dfrac{s(n-1) - s(n+1)}{8} \right\rfloor, \\ x(2n+1) = \dfrac{1}{2}\left(s^{(1)}(n) + d(n) + \left(s^{(1)}(n) \bmod 2\right)\right), \\ x(2n) = s(n) - x(2n+1). \end{cases} \quad (3.29)$$

With respect to Figure 20:

10 **3.1.2.3 Step 2002: Y-direction – High Forward Step**

$$\begin{cases} s(n) = \left\lfloor \dfrac{x(2n) + x(2n+1)}{2} \right\rfloor, \\ d^{(1)}(n) = x(2n+1) - x(2n), \\ d^{(2)}(n) = d^{(1)}(n) + \left\lfloor \dfrac{s(n-1) - s(n+1)}{4} \right\rfloor, \\ d(n) = \left\lfloor \dfrac{d^{(2)}(n)}{2} \right\rfloor, \\ HalfBit(n) = d^{(2)}(n) \bmod 2. \end{cases} \quad (3.30)$$

Inverse Step

$$\begin{cases} d^{(1)}(n) = 2d(n) + HalfBit(n) - \left\lfloor \dfrac{s(n-1) - s(n+1)}{4} \right\rfloor, \\ x(2n) = s(n) - \left\lfloor \dfrac{d^{(1)}(n)}{2} \right\rfloor, \\ x(2n+1) = d^{(1)}(n) + x(2n). \end{cases} \quad (3.31)$$

15

We now compute the approximation error probabilities of our method, and show that it is significantly smaller. We start with the LL-subband error. Assuming e_i

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is the approximation error of the LL-subband in the i – th resolution (Figure 21), the LL-subband coefficient in the i – th resolution can be written as:

$$ll_i^{(new)}(m,n) = \left\lfloor \frac{x_i^{(new)}(2m+1,2n+1) + x_i^{(new)}(2m,2n+1)}{2} \right\rfloor$$
$$+ \left\lfloor \frac{x_i^{(new)}(2m+1,2n) + x_i^{(new)}(2m,2n)}{2} \right\rfloor$$

$$= \frac{x_i^{(acc.)}(2m+1,2n+1) + e_{i-1}^1 + x_i^{(acc.)}(2m,2n+1) + e_{i-1}^2}{2} + e'$$
$$+ \frac{x_i^{(acc.)}(2m+1,2n) + e_{i-1}^3 + x_i^{(acc.)}(2m,2n) + e_{i-1}^4}{2} + e''$$

$$= \frac{x_i^{(acc.)}(2m+1,2n+1) + x_i^{(acc.)}(2m,2n+1) + x_i^{(acc.)}(2m+1,2n) + x_i^{(acc.)}(2m,2n)}{2}$$
$$+ \frac{e_{i-1}^1 + e_{i-1}^2 + e_{i-1}^3 + e_{i-1}^4}{2} + e' + e'' =$$

$$= ll_i^{(acc.)}(m,n) + e_i, \tag{3.32}$$

where

- $ll_i^{(new)}(m,n)$ is the new transform LL-subband coefficient (i-th resolution).
- e_{i-1}^k for $1 \le k \le 4$, are identical random variable representing the error from the previous level.
- e' and e'' are random variables with P.D.F. defined in (3.6).
- $x_i^{(acc.)}(m,n)$ is the i-th resolution image coefficient using (3.1) as the 1D Wavelet step.
- $ll_i^{(acc.)}(m,n)$ is the normalized ($L_2 - norm$) LL-subband coefficient resulting from the i-th 2D transform step using (3.1) as the 1D Wavelet step (see Figure 21).
- $e_i = \frac{e_{i-1}^1 + e_{i-1}^2 + e_{i-1}^3 + e_{i-1}^4}{2} + e' + e''$.

Consequently

$$E(e_i) = \frac{4E(e_{i-1})}{2} + \left(-\frac{1}{4}\right) + \left(-\frac{1}{4}\right) = 2E(e_{i-1}) - \frac{1}{2}, \tag{3.33}$$

$$Var(e_i) = \frac{4Var(e_{i-1})}{4} + \frac{1}{16} + \frac{1}{16} = Var(e_{i-1}) + \frac{1}{8}. \tag{3.34}$$

By knowing that $e_{-1} = 0$ we get

$$\begin{cases} \mathrm{E}\left(e_i\right) = \dfrac{1}{2} - 2^i, \\ \mathrm{Var}\left(e_i\right) = \dfrac{i+1}{8}. \end{cases} \qquad (3.35)$$

Now we can easily evaluate the approximation error of the 3 other subbands:

$$lh_i^{(\mathrm{new})}\left(m,n\right)$$

$$= 2\left| \frac{\left\lfloor \dfrac{x_i^{(\mathrm{new})}\left(2m+1,2n+1\right)+x_i^{(\mathrm{new})}\left(2m,2n+1\right)}{2} \right\rfloor - \left\lfloor \dfrac{x_i^{(\mathrm{new})}\left(2m+1,2n\right)+x_i^{(\mathrm{new})}\left(2m,2n\right)}{2} \right\rfloor}{2} \right| \qquad (3.36)$$

$$= \frac{\left(x_i^{(\mathrm{acc.})}\left(2m+1,2n+1\right)+x_i^{(\mathrm{acc.})}\left(2m,2n+1\right)\right)-\left(x_i^{(\mathrm{acc.})}\left(2m+1,2n\right)+x_i^{(\mathrm{acc.})}\left(2m,2n\right)\right)}{2}$$

$$+ \frac{e_{i-1}^1 + e_{i-1}^2 - e_{i-1}^3 - e_{i-1}^4}{2} + e' - e'' + 2e'''$$

$$= lh_i^{(\mathrm{acc.})}\left(m,n\right) + e_i^{LH},$$

where

- $lh_i^{(\mathrm{new})}\left(m,n\right)$ is the new transform LH-subband coefficient (i-th resolution).

- $lh_i^{(\mathrm{accurate})}\left(m,n\right)$ is the normalized ($L_2 - norm$) LH-subband coefficient resulting from the i-th 2D transform step using (3.1) as the 1D Wavelet step (see Figure 21).

- e''' is a random variable with P.D.F. defined in (3.6).

- $e_i^{LH} = \dfrac{e_{i-1}^1 + e_{i-1}^2 - e_{i-1}^3 - e_{i-1}^4}{2} + e' - e'' + 2e'''$

- all other symbols are defined like in (3.32).

25

5   Hence

$$E\left(e_i^{LH}\right) = \frac{2E\left(e_{i-1}\right) - 2E\left(e_{i-1}\right)}{2} + \left(-\frac{1}{4}\right) - \left(-\frac{1}{4}\right) + 2 \cdot \left(-\frac{1}{4}\right) = -\frac{1}{2}, \qquad (3.37)$$

$$Var\left(e_i^{LH}\right) = \frac{4Var\left(e_{i-1}\right)}{4} + \frac{1}{16} + \frac{1}{16} + 4 \cdot \frac{1}{16} = \frac{i+3}{8}. \qquad (3.38)$$

Similar estimation can be done for the HL and the HH subbands.

10   The error estimation (for all subbands) are summarized in the following table where the error is the difference between the normalized (in $L_2 - norm$) coefficients according to our new reversible transform and the "mathematical" transform (defined in (3.1)).

	Expectation	Variance	Std
LL_i - error	$\frac{1}{2} - 2^i$	$\frac{i+1}{8}$	$\sqrt{\frac{i+1}{8}}$
LH_i - error	$-\frac{1}{2}$	$\frac{i+3}{8}$	$\sqrt{\frac{i+3}{8}}$
HL_i - error	$-\frac{1}{4}$	$\frac{i}{8} + \frac{1}{16}$	$\sqrt{\frac{i}{8} + \frac{1}{16}}$
HH_i - error	$-\frac{1}{4}$	$\frac{i}{8} + \frac{1}{16}$	$\sqrt{\frac{i}{8} + \frac{1}{16}}$

15

Table 3 - Normalized (in $L_2 - norm$) approximation errors of the Wavelet coefficients at resolution $i \geq 0$ ($i = 0$ is the highest resolution) using the proposed reversible Haar transform. The result for the LL-subband is valid for the proposed reversible (1,3)

20   **transform also.**

The meaning of this result is that in a low bit-rate, where only large coefficients are encoded, this error is negligible.

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5   4.   IMAGING PROTOCOL AND DISTRIBUTED DATABASE

Dividing the data into tiles and bit-planes

For the purpose of efficient rendering the coefficients may be sub-divided into

tiles. The tiles of this invention differ from previous art as shown in Figure 12. As in

10   the lossy algorithm, here also the subband tiles are further decomposed to subband data

blocks. Each data block of lossless subband tile (Figure 12) will have a 4D coordinate

$$(t\_x, t\_y, t\_resolution, t\_bitPlane)$$

where $0 \leq t\_bitPlane \leq maxBitPlane(t\_resolution)$.

Each data block contains the following data in encoded format:

15   1. For **t_bitPlane** ≥ 2:

    a. A list of the indices of all subband coefficients whose absolute value is

        in the range $\left[2^{t\_bitPlane-1}, 2^{t\_bitPlane} \right)$.

    b. The sign of all the coefficients of a.

    c. For t_bitPlane >2, an additional precision bit for any coefficient that

20   belongs to the current bit plane or any higher bit plane.

2. For **t_bitPlane =1**, which we call the **"least significant bit plane"**:

    a. A list of the indices of HL-subband and HH-subband coefficients whose

        absolute value belongs to the set $\{-1, 0, 1\}$.

    b. A "zero flag" for each coefficient of a, which indicates if the coefficient

25   is equal to zero or not.

    c. The sign of all the coefficients of a, whose "zero flag" is false.

    d. The LSB of the HL-subband and HH-subband coefficients that belong to

        higher bit plane.

    Remark:

30   Since the LH-subband contains only even coefficients, their LSB must be

    zero and is not coded.

3. For **t_bitPlane =0**, which we call the **"half bit plane"**, a matrix of

    $\left(\dfrac{tileLength}{2} \right)^2$ bits associated with HH-subband coefficients as their last "half

    bit" (See (3.24) or (3.30)).

5. THE PROGRESSIVE SUBBAND CODING ALGORITHM

5.1 The encoding algorithm

The encoding algorithm of the present invention is performed at the server 120. In the present imaging system this rather time consuming task is performed locally in near real-time for a ROI, and not on the full image. The encoding algorithm is described for images with a single color component, such as grayscale images, but of course may also be applied to images with multiple color components. The straightforward generalization for an arbitrary number of components will be explained later.

The lossless algorithm receive as input the following parameters:

Variable	Meaning
coef	Matrix of subband coefficients, containing $3 \times \left(\dfrac{tileLength}{2} \right)^2$ coefficients
HalfBit	Matrix of bits containing $\left(\dfrac{tileLength}{2} \right)^2$ bits.

Table 4 - Lossless Encoding Algorithm Input Parameters

The coding strategy is similar in some sense to that described in A. Said and W. Pearlman, "A new, fast and efficient image codec based on set partitioning", IEEE Trans. Circuits and Systems for video Tech., Vol. 6, No. 3, pp. 243-250, 1996, but the preferred embodiment uses no "Zero Tree" data. For all the data blocks with t_bitPlane ≥ 2, we use the lossy encoding algorithm described in previous art with the parameters:

- *coef* := *coef* (The lossy parameter coef initialized with the lossless parameter coef)
- *equalBinSize* := True

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5 • $\varepsilon_c := 2$

Remark: The lossy algorithm encodes all the bit-plane information for $t\_bitPlane \geq 2$.

For t_bitPlane ≤ 1, i.e. the least significant bit plane (of the lossless algorithm) and the

10 half bit plane, we use a different algorithm described in 5.1.3.

5.1.1 Encoding algorithm initialization

The lossless encoding algorithm initialization is the same as the lossy algorithm of § 4.1.1 in the above-cited Ser. No. 09/386,264, which disclosure is incorporated

15 herein by reference. In order to initialize the encoding algorithm, the following procedure is performed:

1. Assign to each coefficient $coef(x, y)$ its bit plane $b(x, y)$ such that:

$$\left|coef(x, y)\right| \in \left[\varepsilon_c 2^b, \varepsilon_c 2^{b+1}\right)$$

20

2. Compute the maximum bit plane over all such coefficients:

$$maxBitPlane(tile) = \max_{x,y}\left(b(x, y)\right)$$

25 3. Write the value of $maxBitPlane(tile)$ using one byte as the header of the data block:

$$\left(t\_x, t\_y, t\_resolution, maxBitPlane(t\_resolution)\right)$$

4. Initialize all the coefficients as members of their corresponding *Type*16

30 group.

5. Initialize a list of significant coefficients to be empty.

5   6.   Initialize a coefficient approximation matrix *coef* as zero.

5.1.2 The outer loop

The outer loop of the encoding algorithm scans the bit planes from $b = maxBitPlane(tile)$ to $b = 0$. The output of each such bit plane scan is the subband data block. Since the last stage of the encoding algorithm is arithmetic encoding of given symbols, at the beginning of each scan the arithmetic encoding output module is redirected to the storage area allocated for the particular data block. Once the bit plane scan is finished and the data block has been encoded, the output stream is closed and the bit plane b is decremented. After the outer loop is finished the following stages are performed:

1. **Least significant bit plane** is encoded (t_bitPlane = 1).
2. **Half bit plane** is encoded (t_bitPlane = 0).

The output of the least significant bit plane scan is the data block (Figure 14):

(t_x, t_y, t_resolution, t_bitPlane=1).

The half bit plane data block is:

(t_x, t_y, t_resolution, t_bitPlane=0).

5.1.3 Bit-plane scan

For t_bitPlane ≥ 2, the framework of the bit plane scan is described in Figure 24, while the pseudo code is given in the above-cited Ser. No. 09/386,264, which disclosure is incorporated herein by reference. The scan, for a given level b $(b \geq 2)$, encodes all of the coefficients' data corresponding to the absolute value interval $[2^{b-1}, 2^b)$.

Remark: The encoder method isLastBitPlane() is associated to the t_bitPlane = 2.

5   For the least significant bit plane, a pseudo code is described in Figure 15, while a flow
chart is described in Figure 22.

Explanations to the least significant bit encoding algorithm:

1. The coefficients scanning procedure, i.e. moreCoef() procedure in Figure 15 or
"More coefficients?" in Figure 22 includes all the coefficients belong to the HH and

10   the HL-subband (Figure 14). The LH-subband is skipped, since the least significant
bit of each coefficient in it is zero (see remark 2 for (3.22)).

2. The procedure isCoefReporeted() ("Is coefficient reported?" in the flow chart)
returns false if the coefficient is one of $\{-1,0,1\}$, i.e. in all higher bit plane scans it
was insignificant, otherwise it returns true.

15   3. The procedure isCoefExactZero() ("Coefficient is zero?" in the flow chart) returns
true iff the coefficient is zero.

4. The procedure getCoefSign() returns the coefficient's sign.

For the half bit plane, a pseudo code is described in Figure 16.

20

5.2 The decoding algorithm

Obviously, this algorithm is a reversed step of the encoding algorithm of section

25   5.1, performed in the server 120. The client computer 110 during the progressive
rendering operation performs the decoding algorithm. Similar to the encoding
algorithm, the decoding algorithm is described for an image with one component (such
as a grayscale image), but of course could also be used with an image with more than
one component. The input parameters to the lossless algorithm are given below:

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Variable	Meaning
coef	Empty matrix of subband coefficients to be filled by the decoding algorithm.
HalfBit	Matrix of bits containing $\left(\dfrac{tileLength}{2}\right)^2$ bits.

Table 5 - Lossless decoding algorithm input parameters

For all the data blocks with t_bitPlane ≥ 2, a "lossy" decoding algorithm is utilized. The input parameters for the lossy algorithm are:

- *coef* $:=$ *coef*

- *equalBinSize* $:=$ True

- $\varepsilon_c := 2$

5.2.1 Decoding algorithm initialization

1. Assign to each coefficient $Coef(x,y)$ the value zero.

2. Assign to each bit belongs to the HalfBit matrix the value zero.

3. Initialize all the coefficients as members of their corresponding *Type*16 group.

4. Initialize the list of significant coefficients to be empty.

5. If the "first" data block

$$\left(t\_x, t\_y, t\_resolution, maxBitPlane(t\_resolution)\right)$$

is available at the client, read the first byte, which is the value of $maxBitPlane(tile)$.

5.2.2 The outer loop

Upon the completion of the outer loop in 5.1.2 the following stages are preformed:

1. The decoding algorithm scans the least significant bit plane. The input to this stage is encoded data block (t_x, t_y, t_resolution, LeastSiginificant_bitPlane).

2. The decoding algorithm scans the half bit plane. The input to this stage is encoded data block (t_x, t_y, t_resolution, Half_bitPlane).

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5   **5.2.3 Bit plane scan**

The preferred embodiment follows the lossy prior art of the above-cited Ser. No. 09/386,264, which disclosure is incorporated herein by reference, for $t\_bitPlane \geq 2$. The scan, for a given level b, decodes all of the coefficients' data corresponding to the

10   absolute value interval $\left[\varepsilon 2^b, \varepsilon 2^{b+1}\right)$. Pseudo codes of the least significant bit plane scan and half bit plane scan are described in Figure 16.

Explanations to the pseudo code:

1. The decoder's method **moreCoef()** scans all the coefficients in the HH, HL **and LH** subband. But, since the LH-subband is skipped in the encoding algorithm, we don't

15   call to **decodeSymbol()** for its coefficients. Instead of this, we update their least significant bit as zero.

2. Recall that LH-subband coefficients that have not been reported until the least significant bit-plane must be zero since they are known to be even.

20

6.   CLIENT WORKFLOW

With reference to Figure 4, we describe the workflow at the client unit 110. Any new ROI generated by the user's action such as a zoom-in, a scroll, or a luminance tuning invokes in step 401 a call from the GUI interface to the client imaging module

25   with the new ROI view parameters. The client imaging module then computes in step 402 which data blocks are required for the rendering of the ROI and checks if these data blocks are available in the client cache. If not, their coordinate is added to a request list ordered according to some progressive mode. The request list is then encoded in step 403 and sent to the server. The server responds to the request by

30   sending back to the client a stream of data blocks, in the order in which they were requested. In step 404 the client inserts them to their appropriate location in the distributed database. At various points in time step 405, a rendering of the ROI, is invoked. Naturally, the rendering operation is progressive and is able to use only the currently available data in the client's database.

35

5   **6.1 Step 401:Receiving the ROI Parameters**

The imaging module on the client computer 120 receives from the GUI interface module view parameters detailed in Table 5. These parameters are used to generate a request list for the ROI. The same parameters are used to render the ROI.

10

Table 5

Variable	Meaning
worldPolygon	A 2D polygon in the original image coordinate system
scale	The view resolution. $0 < scale < 1$ implies a low view resolution, $scale = 1$ original resolution and $scale > 1$ a higher than original view resolution
deviceDepth	A number in the set $\{8,16,24\}$ representing the depth of the output device (screen, printer)
viewQuality	A quality factor in the range $[1,7]$ where 1 implies very low quality and 7 implies lossless quality
luminanceMap	If active: a curve defining a mapping of medical images with more than 8 bits (typically 10,12,16 bits) per grayscale values to an 8 bit screen
progressiveMo	One of: Progressive By Accuracy, Progressive By Resolution, Progressive by Spatial Order

15

The basic parameters of a ROI are *worldPolygon* and *scale* which determine uniquely the ROI view. If the ROI is to be rendered onto a viewing device with limited resolution, then a *worldPolygon* containing a large portion of the image will be coupled

20   by a small *scale*. In the case where the rendering is done by a printer, the ROI could be a strip of a proof resolution of the original image that has arrived from the server computer 120. This strip is rendered in parallel to the transmission, such that the printing process will terminate with the end of transmission. The other view parameters determine the way in which the view will be rendered. The parameters

25   *deviceDepth* and *viewQuality* determine the quality of the rendering operation. In

5   cases the viewing device is of low resolution or the user sets the quality parameter to a

lower quality, the transfer size can be reduced significantly.

The parameter *luminanceMap* is typically used in medical imaging for

grayscale images that are of higher resolution than the viewing device. Typically,

10   screens display grayscale images using 8 bits, while medical images sometimes

represent each pixel using 16 bits. Thus, it is necessary to map the bigger pixel range

to the smaller range of $[0, 255]$.

Lastly, the parameter *progressiveMode* determines the order in which data

blocks should be transmitted from the server 120. The "Progressive By Accuracy"

15   mode is the best mode for viewing in low bandwidth environments. "Progressive By

Resolution" mode is easier to implement since it does not require the more

sophisticated accuracy (bit plane) management and therefore is commonly found in

other systems. The superiority of the "progressive by accuracy" mode can be

mathematically proven by showing the superiority of "non-linear approximation" over

20   "linear approximation" for the class of real-life images. See, e.g., R. A. DeVore,

"Nonlinear approximation", Acta Numerica, pp. 51-150, 1998.

The "Progressive by Spatial Order" mode is designed, for example, for a "print

on demand" feature where the ROI is actually a low resolution "proof print" of a high

resolution graphic art work. In this mode the image data is ordered and received in a

25   top to bottom order, such that printing can be done in parallel to the transmission.

However, since lossless compression is mostly required in medical images

transmission, where typically more than 8 bits images are used, we amplify our

discussion on the curve (*luminanceMap* hereinabove) which defines the mapping from

the original image gray scale range (typically 10,12,16 bits) to an 8-bit screen. Further

30   more, in medical images viewing, regardless of the original image depth, mapping is

required in order to control the brightness and contrast of the image.

6.1.1 Luminance mapping

Mapping from original image depth (e.g. 10,12,16 bits) to screen depth

35   (typically 8-bits), is defined by a monotonic function (Figure 17):

35

$$f : \left[0, 2^{original\_image\_depth} - 1 \right] \rightarrow \left[0, 2^{screen\_depth} - 1 \right]. \tag{6.1}$$

The curve influences not only the mapping, i.e. the drawing to the screen, but also the request from the server. To understand that, let us concentrate in the maximal gradient of the curve (Figure 17). In a lossy mode, the request is created such that the image approximation in the client side is close enough to the original image, i.e., the RMS (Root Mean Square Error) is visually negligible. When a curve (mapping function) is applied, the RMS can be increased or reduced. The maximal RMS increasing factor depends on the maximal gradient of the curve as follows:

$$RMS\_increasing\_factor = \frac{RMS\left(f(I), f(\hat{I})\right)}{RMS(I, \hat{I})} \leq \max(f'), \tag{6.2}$$

where
- I is the original image
- $\hat{I}$ is the approximated image
- f is the mapping function
- $RMS(I_1, I_2) = \dfrac{\left\| I_1 - I_2 \right\|_{L_2}}{\text{Image\_size}}$
- $\max(f')$ is the maximal gradient of the curve.

We consider the worst case of the RMS increasing factor i.e.:

$$RMS\_increasing\_factor = Maximal\_gradient = \max(f')$$

If the RMS increasing factor is greater than 1, it means that the "new RMS" may be greater than we consider as visually negligible error. Thus, the request list should be increase (more bit-planes should be requested from the server) in order to improve the approximation accuracy. Conversely, if the RMS increasing factor is smaller than 1, the request listing can be reduced. The exact specification of this is given in the following section.

6.2 Step 402: Creating the request list

In step 402 using the ROI view parameters, the client imaging module at the client computer 110 calculates data blocks request list ordered according to the *progressiveMode*. Given the parameters *worldPolygon* and *Scale*, it may be determined which subband tiles in the "frequency domain" participate in the

36

5   reconstruction of the ROI in the "time domain". These tiles contain all the coefficients

that are required for an "Inverse Subband/Wavelet Transform" (IWT) step that

produces the ROI. First, the parameter $dyadicResolution(ROI)$ is computed, which is

the lowest possible dyadic resolution higher than the resolution of the ROI. Any

subband tiles of a higher resolution than $dyadicResolution(ROI)$ do not participate in

10   the rendering operation. Their associated data blocks are therefore not requested, since

they are visually insignificant for the rendering of the ROI. If $scale \geq 1$, then the

highest resolution subband tiles are required. If $scale \leq 2^{1-numberOfResolutions}$ then only the

lowest resolution tile is required. For any other value of $scale$ we perform the

mapping described below in Table 6.

15

Table 6

scale	highestSubbandResolution
$scale \leq 2^{1-numberOfResolutions}$	1
$2^{1-numberOfResolutions} < scale \leq 1$	$numberOfResolutions - \lfloor -\log_2(scale) \rfloor$
$scale > 1$	$numberOfResolutions$

20   Once it has been determined which subband tiles participate in the rendering of

the ROI, it is necessary to find which of their data blocks are visually significant and in

what order they should be requested. Using well known rate/distortion rules from the

field of image coding (such as is described in S. Mallat and F. Falzon, "Understanding

image transform codes", Proc. SPIE Aerospace Conf., 1997), it is not too difficult to

25   determine an optimal order in which the data blocks should be ordered by the client

imaging module (and thus delivered by the server 120). This optimal order is described

in steps 301-310 of Figure 3 for the "Progressive By Accuracy" mode. The underlying

mathematical principal behind this approach is "Non-Linear Approximation".

First, the subband coefficients with largest absolute values are requested since

30   they represent the most visually significant data such as strong edges in the image.

Notice that high resolution coefficients with large absolute value are requested before

low resolution coefficients with smaller absolute value. Within each given layer of

precision (bit plane) the order of request is according to resolution; low resolution

5   coefficients are requested first and the coefficients of *highestSubbandResolution* are

requested last.

The main difficulty of this step is this: Assume a subband tile is required for the

rendering of the ROI. This means that $t\_resolution \leq dyadicResolution(ROI)$ and the

tile is required in the IWT procedure that reconstructs the ROI. It must be understood

10   which of the data blocks associated with the subband tile represent visually

insignificant data and thus should not be requested. Sending all of the associated data

blocks will not affect the quality of the progressive rendering. However, in many cases

transmitting the "tail" of data blocks associated with high precision is unnecessary

since it will be visually insignificant. In such a case, the user will see that the

15   transmission of the ROI from the server 120 is still in progress, yet the progressive

rendering of the ROI seems to no longer to change the displayed image.

Additionally, the influence of the luminance mapping on the accuracy level of

the requested data block as described below. Supposing for some t_x, t_y and

$t\_resolution$, the set

20    $$\left\{ (t\_x, t\_y, t\_resolution, t\_bitPlane) \middle| T \leq t\_bitPlane \leq maxPlaneBit(t\_resolution) \right\}$$

is requested where T is the minimal bit plane required to the current view. Here, where

the luminance mapping is taken in account, the value of T might be increased or

decreased.

25   The number of bit planes reduced (added) from the request list is

$$\left\lfloor \log_2 \left(\frac{1}{Maximal\_gradient} \right) \right\rfloor.$$

I.e., for those t_x, t_y and $t\_resolution$ mentioned before, the following set is

requested:

$$\left\{ (t\_x, t\_y, t\_resolution, t\_bitPlane) \middle| T' \leq t\_bitPlane \leq maxPlaneBit(t\_resolution) \right\}$$

30   where

$$T' = T + \left\lfloor \log_2 \left(\frac{1}{Maximal\_gradient} \right) \right\rfloor.$$

Examples:

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5   1. Given

- Image depth of 12-bit

- Screen depth of 8-bit

- Linear luminance mapping, I.e., *Maximal gradient* $=\dfrac{2^8}{2^{12}}=2^{-4}$.

The number of bit planes reduced from the request list is:

$$\left\lfloor \log_2\left(\frac{1}{Maximal\ gradient}\right)\right\rfloor = \left\lfloor \log_2\left(\frac{1}{2^{-4}}\right)\right\rfloor = 4.$$

10

2. Given a luminance mapping with *Maximal gradient* $=2$

The number of bit planes reduced from the request list is:

$$\left\lfloor \log_2\left(\frac{1}{Maximal\ gradient}\right)\right\rfloor = \left\lfloor \log_2\left(\frac{1}{2}\right)\right\rfloor = -1.$$

Thus one bit plane is added to the original set.

15

6.3 Step 403: Encoding the request list

The client imaging module in the client computer 110 encodes the request list into a request stream that is sent to the server computer 120 via the communication

20   network 130 (Figure 1). The request list encoding algorithm is a simple rectangle-based procedure. The heuristics of the algorithm is that the requested data block usually can be grouped into data block rectangles. From the request list of data blocks indexed by the encoding algorithm computes structures of the type

25    $$\left\{\left(t\_x, t\_y, t\_resolution, t\_bitPlane\right), n_x, n_y\right\}, \quad n_x, n_y \geq 1$$

(1.3)

Each such structure represents the $n_x \times n_y$ data blocks

$$\left\{\left(t\_x+i, t\_y+j, t\_resolution, t\_bitPlane\right)\right\}, \qquad 0 \leq i < n_x, 0 \leq j < n_y$$

30

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5   The encoding algorithm attempts to create the shortest possible list of structures, collecting the data blocks to the largest possible rectangles can do this. It is important to note that the algorithm ensures the order of data blocks in the request list is not changed, since the server 120 will respond to the request stream by transmitting data blocks in the order in which they were requested. A good example of when this works

10   well is when a user zooms in into a ROI at a high resolution that was never viewed before. In such a case the request list might be composed of hundreds of requested data blocks, but they will be collected to one (x,y) rectangle for each pair $(t\_resolution, t\_bitPlane)$.

15

6.4 Step 404: Receiving the data blocks

The client computer 110 upon receiving from the server computer 120 an encoded stream containing data blocks, decodes the stream and inserts the data blocks into their appropriate location in the distributed database using their ID as a key. The

20   simple decoding algorithm performed here is a reversed step of the encoding infra. Since the client 110 is aware of the order of the data blocks in the encoded stream, only the size of each data block need be reported along with the actual data. In case the server 120 informs of an empty data block, the receiving module marks the appropriate slot in the database as existing but empty.

25   Recall that the subband tile associated with each data block is denoted by the first three coordinates of the four coordinates of a data block $(t\_x, t\_y, t\_resolution)$. From the subband tile's coordinates the dimensions are calculated of the area of visual significance; that is, the portion of the ROI that is affected by the subband tile. Assume that each subband tile is of length *tileLength* and that the wavelet basis used has a

30   maximal filter size *maxFilterSize*, then defining $hFilterSize := \lceil maxFilterSize/2 \rceil$ and $factor := numberOfResolutions - t\_resolution + 1$, we have that the dimensions of the affected region of the ROI (in the original image's coordinate system) are

$$\left[t\_x \times tileLength^{factor} - hFilterSize^{factor}, (t\_x+1) \times tileLength^{factor} + hFilterSize^{factor}\right] \times$$
$$\left[t\_y \times tileLength^{factor} - hFilterSize^{factor}, (t\_y+1) \times tileLength^{factor} + hFilterSize^{factor}\right]$$

These dimensions are merged into the next rendering operation's region. The rendering region is used to efficiently render only the updated portion of the ROI.

6.5 Progressive Rendering

During the transmission of ROI data from the server to the client, the client performs rendering operations of the ROI. To ensure that these rendering tasks do not interrupt the transfer, the client runs two program threads: communications and rendering. The rendering thread runs in the background and draws into a pre-allocated "off-screen" buffer. Only then does the client use device and system dependant tools to output the visual information from the "off-screen" to the rendering device such as the screen or printer.

The rendering algorithm performs reconstruction of the ROI at the highest possible quality based on the available data at the client. That is, data that was previously cached or data that "just" arrived from the server. For efficiency, the progressive rendering is performed only for the portion of the ROI that is affected by newly arrived data. Specifically, data that arrived after the previous rendering task began. This "updated region" is obtained using the method of step 404 described in §6.4.

The parameters of the rendering algorithm are composed of two sets:

1. The ROI parameters described in Table 3.
2. The parameters transmitted from the server explained in Table 5 , with the exception of the *jumpSize* parameter, which is a "server only" parameter.

The rendering algorithm computes pixels at the dyadic resolution *dyadicRresolution(ROI)*. Recall that this is the lowest possible dyadic resolution that is higher than the resolution of the ROI. The obtained image is then resized to the correct

41

5 resolution. Using a tiling of the multiresolution representation of the ROI, the steps of the algorithm are performed on a tile by tile basis as described in Figure 11. Since the tiles' length are *tileLength*, which is typically chosen as 64, the rendering algorithm is memory efficient.

10 **6.5.1 The rendering rate**

As ROI data is transmitted to the client 110, the rendering algorithm is performed at certain time intervals of a few seconds. At each point in time, only one rendering task is performed for any given displayed image. To ensure that progressive rendering does not become a bottleneck, two rates are measured: the data block transfer

15 rate and the ROI rendering speed. If it predicted that the transfer will be finished before a rendering task, a small delay is inserted, such that rendering will be performed after all the data arrives. Therefore, in a slow network scenario (as the Internet often is), for almost the entire progressive rendering tasks, no delay is inserted. With the arrival of every few kilobytes of data, containing the information of a few data blocks,

20 a rendering task visualizes the ROI at the best possible quality. In such a case the user is aware that the bottleneck of the ROI rendering is the slow network and has the option to accept the current rendering as a good enough approximation of the image and not wait for all the data to arrive.

25 **6.5.2 Memory constraint subband data structure**

This data-structure is required to efficiently store subband coefficients, in memory, during the rendering algorithm. This is required since the coefficients are represented in long integer (lossless coding mode) or floating-point (lossy coding

30 mode) precision which typically require more memory than pixel representation (1 byte). In lossy mode, the coefficients at the client side 110 are represented using floating-point representation, even if they were computed at the server side 120 using an integer implementation. This will minimize round-off errors.

At the beginning of the rendering algorithm, coefficient and pixel memory strips

35 are initialized. $dyadicWidth(ROI)$ may be denoted as the width of the projection of the

42

5   ROI onto the resolution $dyadicResolution(ROI)$. For each component and resolution

$1 < j \le dyadic\,Re\,solution(ROI)$, <u>four</u> subband strips are allocated for the four types of

subband coefficients: hl, lh, hh <u>and *HalfBit*</u>. The coefficient strips are allocated with

dimensions

$$\left[2^{j-dyadicResolution(ROI)-1} \times dyadicWidth(ROI), \frac{3}{2} \times tileLength + \frac{maxFilterSize}{2} \right]$$

For each component and resolution $1 \le j < dyadicResolution$ a pixel strip is

allocated with dimensions

$$\left[2^{j-dyadicResolution(ROI)} \times dyadicWidth(ROI), tileLength + \frac{maxFilterSize}{2} \right]$$

15

Beginning with the lowest resolution 1, the algorithm proceeds with a recursive

multiresolution march from the top of the ROI to bottom (y direction. Referring to

Figures 10 and 11, in step 1101, the multiresolution strips are filled with sub-tiles of

coefficients 1050 decoded from the database or read from the memory cache. From the

20   coefficients we obtain multiresolution pixels 1051 using an inverse subband transform

step 1102 (shown in further detail in Figure 10). Each time a tile of pixels at

resolutions $j < dyadicResolution(ROI)$ is reconstructed, it is written into the pixel strip

at the resolution j. Each time a tile of pixels at the highest resolution

$dyadicResolution(ROI)$ is reconstructed, it is fed into the inverse color transform and

25   resizing steps 1103, 1104.

6.5.3 Step 1101: decoding and memory caching

The subband coefficients data structure described previously in section 6.5.2 is

30   filled on a tile basis. Each such subband tile is obtained by decoding the corresponding

data blocks stored in the database or reading from the memory cache. The memory

cache is used to store coefficients in a simple encoded format. The motivation is this:

the decoding algorithm described previously in section 5.2 is computationally intensive

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5   and thus should be avoided whenever possible. To this end the rendering module uses

a memory cache 111 where subband coefficient are stored in very simple encoded

format which decodes very fast. For each required subband tile, the following

extraction procedure is performed, described in Figure 25, beginning at step2501. In

step2502, if no data blocks are available in the database for the subband tile, its

10   coefficients are set to zero (step2503). In step2504, if the tile's memory cache storage

is updated, namely it stores coefficients in the same precision as in the database, then

the coefficients can be efficiently read from there (step2505). In step2506, the last

possibility is that the database holds the subband tile in higher precision. Then, the tile

is decoded down to the lowest available bit plane using the algorithm previously

15   described in section 5.2 and the cached representation is replaced with a representation

of this higher precision information.

6.5.4 Step 1102: inverse lossless wavelet transform

This is an inverse step to step 603 performed in the server (see 7.1.5). Following Figure

20   21 we see that four "extended" subband coefficient sub-tiles of length

$tileLength/2 + maxFilterSize$ at the resolution i are read from the coefficient strips data

structure and transformed to a tile of pixels at the next higher resolution using

$losslessWaveletdTransformType(i)$. If $i+1 < dyadicResolution(ROI)$, the tile of pixels

obtained by this step is inserted into the pixel memory strip at the resolution $i+1$. If

25     $i+1 = dyadicResolution(ROI)$ the tile of pixels is processed by the next step of color

transform.

Remark: Recall from 5.1.1 that the "half bits" are initialized as zeros, therefore the

inverse step is well defined even if their "real" value is not available in the client yet.

6.5.5 Step 1103: inverse color transform

This is an inverse step to step 603 performed at the server 120. It is performed

only for tiles of pixels at the resolution *highestSubbandResolution* . At this stage, all of

the pixels of each such tile are in the *outputColorSpace* and so need to be transformed

into a displaying or printing color space. For example, if the original image at the

35   server 120 is a color image in the color space RGB, the pixels obtained by the previous

44

5   step of inverse subband transform are in the compression color space YUV. To convert back from YUV to RGB, we use the inverse step described in Figure 13. If the ROI is at an exact dyadic resolution, then the tile of pixels in the rendering color space is written into the off-screen buffer. Else it is resized in the next step.

10

6.5.6 Step 1104: image resize

In case the resolution of the ROI is not an exact dyadic resolution, the image obtained by the previous step must be re-sized to this resolution. This can be accomplished using operating system imaging functionality. In most cases the
15   operating system's implementation is sub-sampling which produces in many cases an aliasing effect which is visually not pleasing. To provide higher visual quality, the imaging system of the present invention may use the method of linear interpolation, for example described in J. Proakis and D. Manolakis, "Digital signal processing", Prentice Hall, 1996. The output of the interpolation is written to the off-screen buffer. From
20   there it is displayed on the screen using system device dependant methods.

6.5.7 Step 1105: mapping to 8-bit screen

When *luminanceMap* is active mapping to 8-bit screen is performed using the mapping function described in 6.1.1.
25

7.   SERVER WORFLOW

With reference to Figure 5, the operation of the server computer 120 (Figure 1) will now be described. Initially, an uncompressed digital image is stored in, for example, storage 122 of the server computer 120. This uncompressed digital image
30   may be a two-dimensional image, stored with a selected resolution and depth. For example, in the medical field, the uncompressed digital image may be contained in a DICOM file.

Once the client computer 110 requests to view or print a certain image, the server performs the preprocessing step 501. This step is a computation done on data
35   read from the original digital image. The results of the computation are stored in the

5   server cache device 121. After this fast computation a "ready to serve" message is sent

from the server to the client containing basic information on the image.

In step 502, the server receives an encoded stream of requests for data blocks

associated with a ROI that needs to be rendered at the client. The server then decodes

the request stream and extracts the request list.

10   In step 503, the server reads from cache or encodes data block associated with

low resolution portions of the ROI, using the cached result of the preprocessing stage

501.

If the ROI is a high-resolution portion of the image, the server, in step 504,

reads from cache or performs a "local" and efficient version of the preprocessing step

15   501. Specifically, a local portion of the uncompressed image, associated with the ROI,

is read from the storage 122, processed and encoded. In step 505, the data that was

encoded in steps 503-504 is progressively sent to the client in the order it was

requested.

20   **7.1 Step 501: preprocessing**

The preprocessing step is now described with respect to Figure 6. The

preprocessing algorithm's goal is to provide the fastest response to the user's request to

interact with the image. Once this fast computational step is performed, the server is

able to provide efficient "pixel-on-demand" transmission of any client ROI requests

25   that will follow. In most cases the first ROI is a view of the full image at the highest

resolution that "fits" the viewing device. The preprocessing algorithm begins with a

request for an uncompressed image that has not been processed before or has been

processed but the result of this previous computation has been deleted from the cache.

As explained, this unique algorithm replaces the possibly simpler procedure of

30   encoding the full image into some progressive format. This latter technique will

provide a much slower response to the user's initial request then the technique

described below. At the end of the algorithm a "ready to serve ROI of the image"

message is sent to the client containing basic information on the image. While some of

this information, image dimensions, original color space, resolution etc., is available to

35   the user of the client computer, most of this information is "internal" and required by

5   the client to formulate ROI request lists (§6.2) and progressively render (§6.5). Next we

describe in detail the preprocessing algorithm.

7.1.1 Preprocessing Parameters

Table 7

10

Variable	Meaning
losslessMode	If true, preprocessing will prepare data that can be used for lossless transmission.
subbandTransformType	The framework allows the ability to select a different subband transform for each resolution of each image. The technical term is: non-stationary transforms.
numberOfResolutions	The number of resolutions in the Multiresolution structure calculated for the image. Typically, $numberOfResolutions = \log_2\left(\sqrt{ImageSize}\right)$.
jumpSize	A number in the range $\left[0, numberOfResolutions - 1\right]$. The preprocessing stage computes only the top lower part of the image's multiresolution pyramid of the size $numberOfResolutions - jumpSize$.
tileLength	Typically = 64. Tradeoff between time and coding performance.
nTilesX(j)	Number of subband tiles in the x direction at the resolution j
nTilesX(j)	Number of subband tiles in the y direction at the resolution j
inputColorSpace	Color space of uncompressed original image.
outputColorSpace	Transmitted color space used as part of the encoding technique.
numberOfComponents	Number of components in *OutputColorSpace*.
threshold(c, j)	A matrix of values used in lossy compression. The subband coefficients of the component c at the resolution j with absolute value below *threshold(c, j)* are considered as (visually) insignificant and set to zero.

Given an input image, the parameters described in Table 7 are computed or

chosen. These parameters are also written into a header sent to the client 110 and are

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5   used during the progressive rendering step 405 (see section 6.5, described previously). The important parameters to select are:

1.   *losslessMode*: In this mode, progressive transmission of images takes place until lossless quality is obtained. Choosing this mode requires the preprocessing algorithm to use certain reversible wavelet transforms, and can slow down the algorithm.

2.   *subbandTransformType* (j): The (dynamic) selection of wavelet basis (as described, for example, in I. Daubechies, "Ten lectures on wavelets", Siam, 1992) is crucial to the performance of the imaging system. The selection can be non-stationary, meaning a different transform for each resolution j. The selection is derived from the following considerations:

a.   Coding performance (in a rate/distortion sense): This is obviously required from any decent subband/wavelet transform.

b.   Approximation of ideal low pass: It is favorable to choose a transform such that low resolutions of the image will be of high visual quality (some filters produce poor quality low resolutions even before any compression takes place).

c.   Fast transform implementation: Can the associated fast transform be implemented using lifting steps (as described, for example, by I. Daubechies and W. Sweldens, "Factoring wavelet transforms into lifting steps", J. Fourier Anal. Appl., Vol. 4, No. 3, pp. 247-269, 1998), using only integer shifts and additions, etc. Some good examples are the Haar and CDF transforms (1,3), (2,2)*** described in I. Daubechies, "Ten lectures on wavelets", Siam, 1992.

d.   Fast low pass implementation: A very important parameter, since together with the parameter *jumpSize*, it determines almost all of the complexity of the algorithm. For example, the

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5

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CDF (1,3) is in this respect the "optimal" transform with three vanishing moments. Since the dual scaling function is the simple B-spline of order 1, its low pass is simple averaging. Thus, the sequence of CDF transforms, using the B-spline of order 1 as the dual scaling function, but with wavelets with increasing number of vanishing moments are in some sense optimal in the present system. They provide a framework for both real time response and good coding efficiency.

e. Lossless mode: If *losslessMode* is true we must choose the filters from a subclass of reversible transforms (see, for example, "Wavelet transforms that map integers to integers", A. Calderbank, I. Daubechies, W. Sweldens, B. L. Yeo, J. Fourier Anal. Appl., 1998).

15

f. Low system I/O: If the network 130 in Figure 1 connecting between the Image residing on the storage 122 and the imaging server 120 is slow, the bottleneck of the preprocessing stage (and the whole imaging system for that fact) might be simply the reading of the original image. In such a case a transform may be chosen with a lazy sub-sampling low pass filter that corresponds to efficient selective reading of the input image. Many interpolating subband transforms with increasing number of vanishing moments can be selected to suit this requirement. However, this choice should be avoided whenever possible, since it conflicts with (a) and (b).

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g. Image type: If the type of the image is known in advance, an appropriate transform can be chosen to increase coding efficiency. For example: Haar wavelet for graphic images, smoother wavelet for real-life images, etc. In the graphic arts field, there are numerous cases of documents composed of low resolution real-life images and high resolution graphic content. In such a case, a non-stationary sequence of transforms may be chosen: Haar for the high resolutions and a smoother basis

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5   starting at the highest resolution of a real-life image part. In case of low system I/O (f), a non-stationary choice of interpolating transforms of different orders is required.

3.   *jumpSize*: This parameter controls the tradeoff between fast response

10   to the user's initial request to interact with the image and response times to subsequent ROI requests. When *jumpSize* is large, the initial response is faster, but each computation of a region of interest with higher resolution than the jump might require processing of a large portion of the original image.

15   4.   *InputColorSpace*: The input color spaces supported in lossless mode are:

       a.   Grayscale: For grayscale images

       b.   RGB

       5.   *outputColorSpace*: The following are color spaces which perform well

20   in coding:

       a.   Grayscale: For grayscale images

       b.   YUV: for viewing color images

       Referring to Table 7 [LP], *losslessMode* is set to true. *Threshold* (c, j) is not in

25   use, since in lossless mode, there is no thresholding. The rest of the variables have the same meaning as in the lossy algorithm.

7.1.2 Memory constraint multiresolution scan data structure

30   Most wavelet coding algorithms have not addressed the problem of memory complexity. Usually the authors have assumed there is sufficient memory such that the image can be transformed in memory from the time domain to a wavelet frequency domain representation. It seems the upcoming JPEG2000 will address this issue, as did its predecessor JPEG. The preprocessing algorithm also requires performing subband

35   transforms on large images, although not always on the full image, and thus requires

5   careful memory management. This means that the memory usage of the algorithm is not of the order of magnitude of the original image, as described in J. M. Shapiro,"An embedded hierarchical image coder using zero-trees of wavelet coefficients", IEEE Trans. Sig. Proc., Vol. 41, No. 12, pp. 3445-3462, 1993.

Given an uncompressed image we allocate the following number of memory

10   strips

$$numberOfComponents \times \left(numberOfResolutions - jumpSize \right)$$

of sizes

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$$\left[2^{-\left(numberOfResolutions - j \right)} imageWidth, 3 \times tileLength / 2 + maxFilterSize \right]$$

for $1 \le j \le numberOfResolutions - jumpSize - 1$ and

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$$\left[imageWidth, tileLength + 2 \times maxFilterSize \right]$$

for $j = numberOfResolutions - jumpSize$

That is, the memory usage is proportional to $2^{-jumpSize} \times imageWidth$. Each such

25   strip stores low-pass coefficients in the color space *outputColorSpace* at various resolutions.

Referring to Figure 6, during the preprocessing stage, the resolutions are scanned simultaneously from start to end in the y direction. For each color component and resolution, the corresponding strip stores low-pass coefficients or pixels at that

30   resolution. The core of the preprocessing algorithm are steps 604-607, where tiles of pixels of length $tileLength + 2 \times maxFilterSize$ are read from the memory strips and handled one at a time. In step 604 the tile is transformed into a tile of length *tileLength* containing two types of coefficient data: subband coefficients and pixels at a lower resolution. The subband coefficients are processed in steps 605-606 and are stored in

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5   the cache. The lower resolution pixels are inserted in step 607 to a lower resolution memory strip. Whenever such a new sub-tile of lower resolution pixels is inserted into a strip, the algorithm's memory management module performs the following check: If the part of the sub-tile exceeds the current virtual boundaries of the strip, the corresponding first lines of the strip are considered unnecessary anymore. Their

10   memory is (virtually) re-allocated for the purpose of storing the sub-tile's new data.

7.1.3 Step 601: Lossless color-transform

    This step is uses the conversion formula described in Figure 13. This step must be performed before step 602, because the lossless color conversion is non-linear.

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7.1.4 Step 602: Lossless Wavelet Low Pass

    The motivation for the low pass step is explained in § 6.1.4 in the above-cited Ser. No. 09/386,264, which disclosure is incorporated herein by reference. In a lossless mode there are a few emphasis that are represented here.

20   In step 602, the low pass filters of the transforms $subbandTransformType(j)$,

$numberOfResolutions - jumpSize < j \leq numberOfResolutions$, are used to obtain a low resolution strip at the resolution $numberOfResolutions - jumpSize$ (as can be seen in Figure 26). Typically, it is required to low pass about $2^{jumpSize}$ lines of the original

25   image 1010 to produce one line of the low resolution image. The low pass calculation is initiated by a read of tiles of pixels from the memory strips performed in step 604. Whenever there is an attempt to read missing low resolution lines, they are computed by low passing the original image and inserted into the memory strip. The insertion over-writes lines are no longer required, such that the algorithm is memory constrained.

30   In the case where a non-linear color transform took place in the previous step 601, the results of that transform are low-passed.

    In the lossless mode of operation, the *jumpSize* parameter defines the number of lossless wavelet low pass steps should be done. A single low pass step is the same for Haar and CDF (1,3) and defined by the following two stages (taken from (3.20) and

35   (3.22):

X-direction: $\quad s(n) = \left\lfloor \dfrac{x(2n) + x(2n+1)}{2} \right\rfloor$

Y-direction: $\quad s(n) = x(2n) + x(2n+1)$.

Namely, in a 2D representation, the low pass step is defined by

$$ll(m,n) = \left\lfloor \frac{x(2m+1, 2n+1) + x(2m, 2n+1)}{2} \right\rfloor + \left\lfloor \frac{x(2m+1, 2n) + x(2m, 2n)}{2} \right\rfloor . \quad (7.1)$$

For $jumpSize = 1$ and $jumpSize = 2$ (other sizes practically are not needed), the server performs these steps efficiently (almost like the lossy algorithm) by a single operation that simulates exactly $jumpSize$ low pass steps defined in (7.1). As noticed from (7.1), the simplicity of the formula makes filters such as Haar and CDF (1,3) "optimal" in the respect of low pass efficiency.

7.1.5 Step 603: Forward lossless Wavelet transform

In Step 603 we perform one step of an efficient local lossless wavelet transform (§3), on a tile of pixels at the resolution $1 \leq j \leq numberOfResolutions - jumpSize$. The type of transform is determined by the parameter $losslessWaveletTransformType(j)$. As described in Figure 1, the transform is performed on an "extended" tile of pixels is of length $tileLength + 2 \times maxFilterSize$ (unless we are at the boundaries), read directly from a multi-resolution strip at the resolution $j+1$. The output of the step is a lossless subband tile composed of wavelet coefficients including Half bit coefficients and low resolution coefficients/pixels. The transform step is efficiently implemented in integers as described in §3.

The subband transform of step 603 outputs three types of data: scaling function (low pass), wavelet (high pass) and Halfbits. The wavelet coefficients are treated in step 604 while the scaling function coefficients are treated in step 605.

Remark: Tiles of pixels which are located on the boundaries sometimes need to be padded by extra rows and/or columns of pixels, such that they will formulate a "full" tile of length $tileLength$.

7.1.6 Step 604: Variable Length encoding and storage

In step 604, the subband coefficients that are calculated in step 603 are variable length encoded and stored in the cache 121. If $maxBitPlane(tile) = 0$ we do not write any data. Else we loop on the coefficient groups $\{coef(2 \times i + x, 2 \times j + y)\}_{x,y=0,1}$. For each such group we first write the group's variable length $length(i,j)$ using $\log_2(maxBitPlane(tile))$ bits. Then for each coefficient in the group we write $length(i,j)+1$ bits representing the coefficient's value. The least significant bit represents the coefficient's sign: if it is 1 then the variable length encoded coefficient is assumed to be negative. The HalfBit subband coefficients are written in one-bit per coefficient.

7.1.7 Step 605: Copying low pass coefficients into the multiresolution strip structure

In step 503, unless *losslessMode* is true, the subband coefficients calculated in step 604 are quantized. This procedure is performed at this time for the following reason: It is required that the coefficients computed in the previous step will be stored in the cache 121. To avoid writing huge amounts of data to the cache, some compression is required. Thus, the quantization step serves as a preparation step for the next variable length encoding step. It is important to point out that the quantization step has no effect on compression results. Namely, the quantization step is synchronized with the encoding algorithm such that the results of the encoding algorithm of quantized and non-quantized coefficients are identical.

A tile of an image component c at the resolution j is quantized using the given threshold $threshold(c,j)$: for each coefficients x, the quantized value is $\lfloor x/threshold(c,j) \rfloor$. It is advantageous to choose the parameters $threshold(c,j)$ to be dyadic such that the quantization can be implemented using integer shifts. The quantization procedure performed on a subband tile is as follows:

1. Initialize $maxBitPlane(tile) = 0$.

2. Loop over each group of four coefficients $\{coef(2 \times i + x, 2 \times j + y)\}_{x,y=0,1}$. For each such group initialize a variable length parameter $length(i,j) = 0$.

3. Quantize each coefficient in the group $coef(2 \times i + x, 2 \times j + y)$ using the appropriate threshold.

4. For each coefficient, update $length(i,j)$ by the bit plane b of the coefficient, where the bit plane is defined by

$$\left| coef(2 \times i + x, 2 \times j + y) \right| \in \left[2^b\, threshold(c,j), 2^{b+1}\, threshold(c,j) \right)$$

5. After processing the group of four coefficients, use the final value of $length(i,j)$ to update $maxBitPlane(tile)$ by

$$maxBitPlane(tile) = max\left(maxBitPlane(tile), length(i,j) \right)$$

6. At the end of the quantization step, store the value $maxBitPlane(tile)$ in the cache 121.

Note that for subband tiles located at the boundaries we can set to zero subband coefficients that are not associated with the actual image, but only with a padded portion. To do this we take into account the amount of padding and the parameter $maxFilterSize$. The motivation for the "removal" of these coefficients is coding efficiency.

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5   **7.2 Step 502: Decoding the request stream**

This is the inverse step of section 6.3. Once the request stream arrives at the server 120, it is decoded back to a data block request list. Each data structure the type representing a group of requested data blocks is converted to the sub-list of these data blocks.

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7.3 Step 503: Encoding low resolution part of ROI

Step 503 is described in Figure 7. It is only performed whenever the data blocks associated with low-resolution subband tile are not available in the server cache 121.

15   Step 701 is the inverse step of step 604 described in §7.1.6. In the preprocessing algorithm subband tiles of lower resolution, that is resolutions lower than *numberOfResolutions − jumpSize*, were stored in the cache using a variable length type algorithm. For such a tile we first need to decode the variable length representation. The algorithm uses the stored value $maxBitPlane(tile)$.

20   1. If $maxBitPlane(tile) = 0$, then all the coefficients are set to zero including the HalfBit subband.

   2. If $maxBitPlane(tile) = 1$, then all the coefficients are set to zero, and the HalfBit subband coefficient are read bit by bit from cache.

   3. Else, as performed in 2, the HalfBit subband coefficient are read bit by bit from
25   cache, and we perform the following simple decoding algorithm:

For each group of four coefficients $\{coef(2 \times i + x, 2 \times j + y)\}_{x,y=0,1}$, we read

$\log_2(maxBitPlane(tile))$ bits representing the variable length of the group.

Assume the variable length is $length(i,j)$. For each of the four coefficients we
30   then read $length(i,j)+1$ bits. The least significant bit represents the sign. The reconstructed coefficient takes the value:

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$$\left(readBits \gg 1\right) \times \begin{cases} -1 & readBits \,\&\, 1 = 1 \\ 1 & readBits \,\&\, 1 = 0 \end{cases}$$

In step 702 we use the encoding algorithm described in §5.1 to encode the requested data blocks associated with the extracted subband tile.

7.4 Step 504: Processing high resolution part of ROI

Step 504 is described in Figure 8. In case we have used $jumpSize > 0$ in step 501 and the resolution of the ROI $> numberOfResolutions\ -jumpSize$, we are sometimes required to perform a local variation of the preprocessing step described in §7.1. Whenever the server receives a request list of data blocks we check the following. If a data block has been previously computed (present in the cache 121) or is associated with a low resolution subband tile data block then it is either simply read from the cache or handled in step 503. Else, the coordinates of the data block are used to find the "minimal" portion of the ROI that needs to be processed. Then, a local version of the preprocessing algorithm is performed for this local portion. The difference here is that step 804 replaces Variable Length coding step 604 of the preprocessing algorithm by the encoding algorithm given in §5.1.

7.5 Step 505: Progressive transmission of ROI

In the final step, the encoded data tiles are sent from the server 120 to the client 110, in the order they were requested. In many cases, data blocks will be empty. For example, for a region of the original image with a constant pixel value, all of the corresponding data blocks will be empty, except for the first one that will contain only one byte with the value zero representing $maxBitPlane(tile) = 0$. For a low activity region of the image, only the last data blocks representing higher accuracy will contain any data. Therefore, to avoid the extra side information, rectangles of empty data blocks are collected and reported to the client 110 under the restriction that they are reported in the order in which they were requested. For blocks containing actual data, only the data block's size in bytes need be reported, since the client 110 already knows which data blocks to expect and in which order.

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5   The present invention has been described in only a few embodiments, and with respect to only a few applications (e.g., commercial printing and medical imaging). Those of ordinary skill in the art will recognize that the teachings of the present invention may be used in a variety of other applications where images are to be transmitted over a communication media